



Pre-recorded sessions:
From 4 December 2020

Live sessions:
10 – 13 December 2020

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MapTree: Recovering Multiple Solutions in the Space of Maps

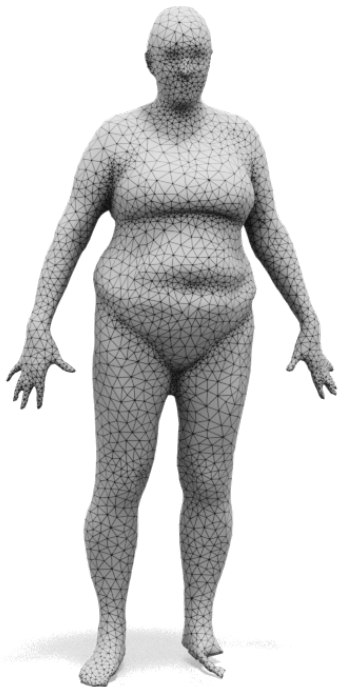
Jing Ren, KAUST

Simone Melzi, Ecole Polytechnique

Maks Ovsjanikov, Ecole Polytechnique

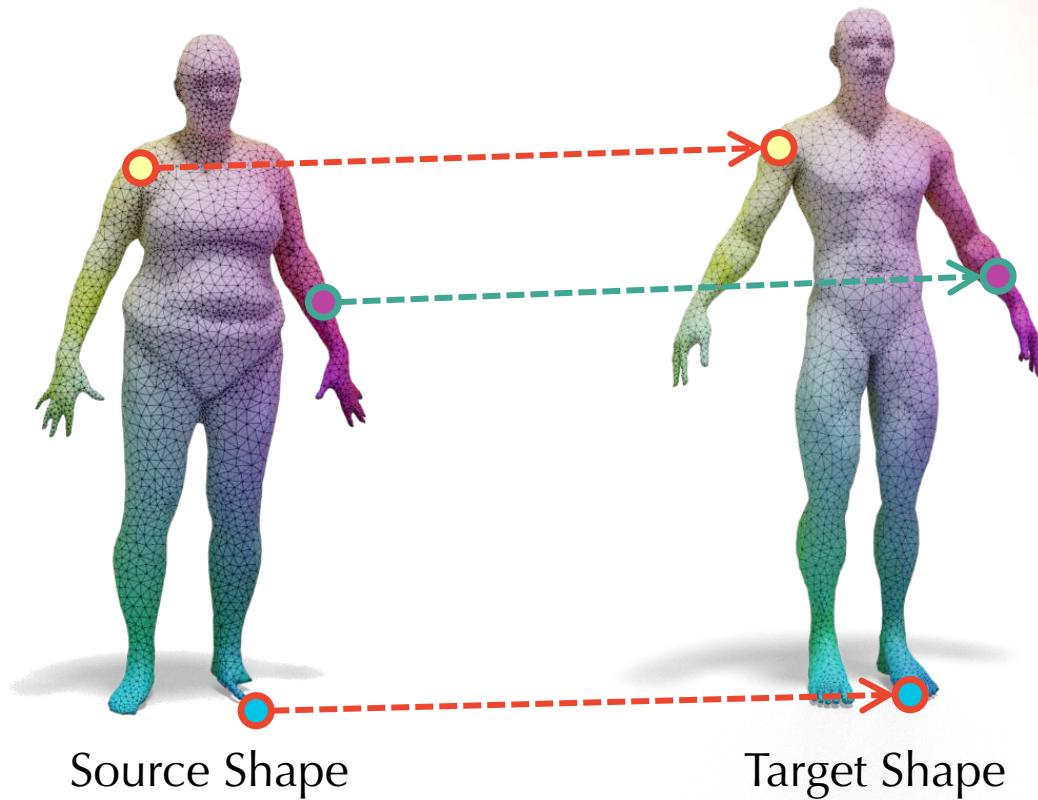
Peter Wonka, KAUST

Map & Correspondences

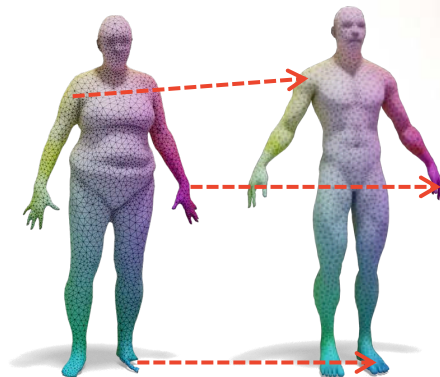


Triangle meshes

Map & Correspondences



- Point-based methods
 - [Bronstein et al. 2006],
 - [Huang et al. 2008]...
- Parametrization-based methods
 - [Lipman and Funkhouser 2009]
 - [Aigerman et al. 2017]...
- Optimal transport
 - [Solomon et al. 2016]
 - [Mandad et al. 2017]...
- Functional maps
 - [Ovsjanikov et al. 2012]
 - [Ezuz and Ben-Chen 2017]...

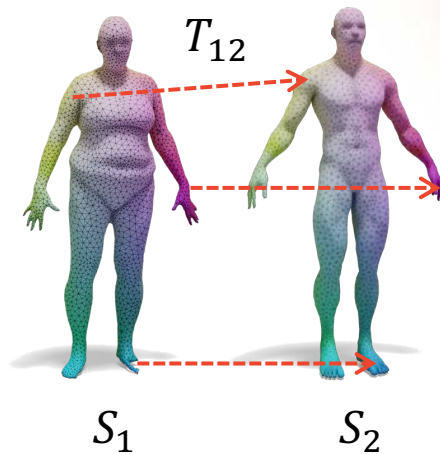


How to compute a map?

Problem Formulation

$$\min_{T_{12}} E(T_{12})$$

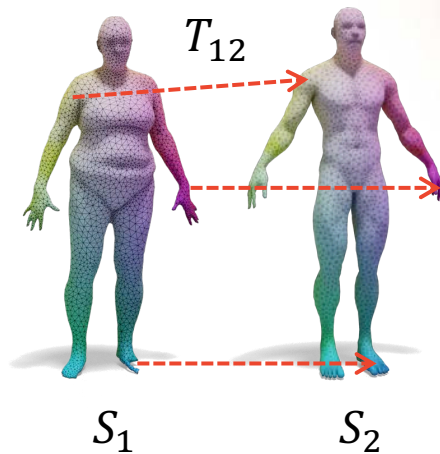
- **Objectives** $E(\cdot)$
 - geodesic distortion
 - Dirichlet energy
 - ...
- **Constraints**
 - bijective map
 - ...



Geodesic distortion

$$E(T_{12}) = \sum_{(v_i, v_j)} \left\| D_1(v_i, v_j) - D_2(T_{12}(v_i), T_{12}(v_j)) \right\|$$

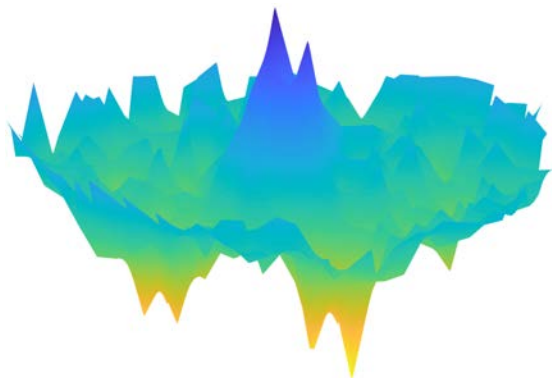
- $D_k(v_i, v_j)$ stores the **geodesic distance** between the two vertices v_i and v_j on shape S_k



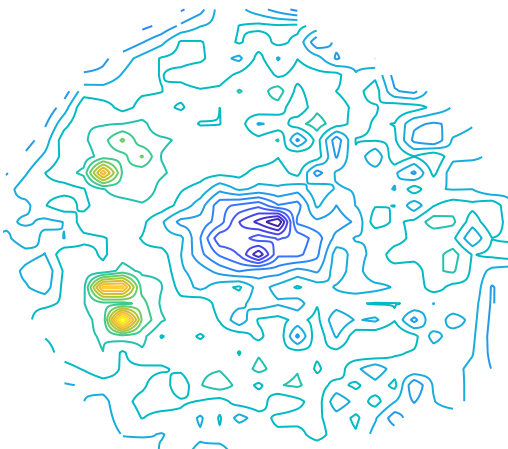
Complicated map space

Geodesic distortion

$$E(T_{12}) = \sum_{(v_i, v_j)} \left\| D_1(v_i, v_j) - D_2(T_{12}(v_i), T_{12}(v_j)) \right\|$$



landscape



contour

Map space

- Discrete
- Not differentiable w.r.t. T_{12}
- Complicated $\mathcal{O}(n^n)$
- Multiple local-minima
-

Avoid undesired local minima?

Existing solutions

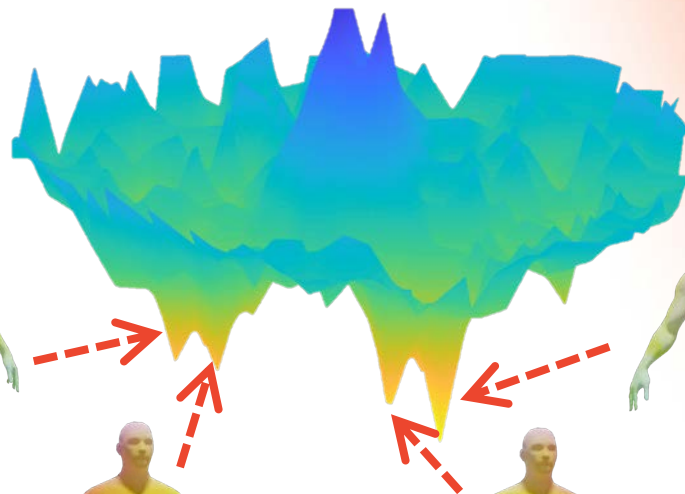
- [Sung and Kim 2013] “Finding the M-best consistent correspondences between 3D symmetric objects”
 - For each vertex, find multiple correspondences candidates to resolve global symmetry ambiguity
- [Sahillioglu and Yemez 2013] “Coarse-to-fine Isometric Shape Correspondence by tracking symmetric flips”
 - Avoid symmetry flip during map computation

Find all meaningful maps!

Landscape of the
geodesic distortion

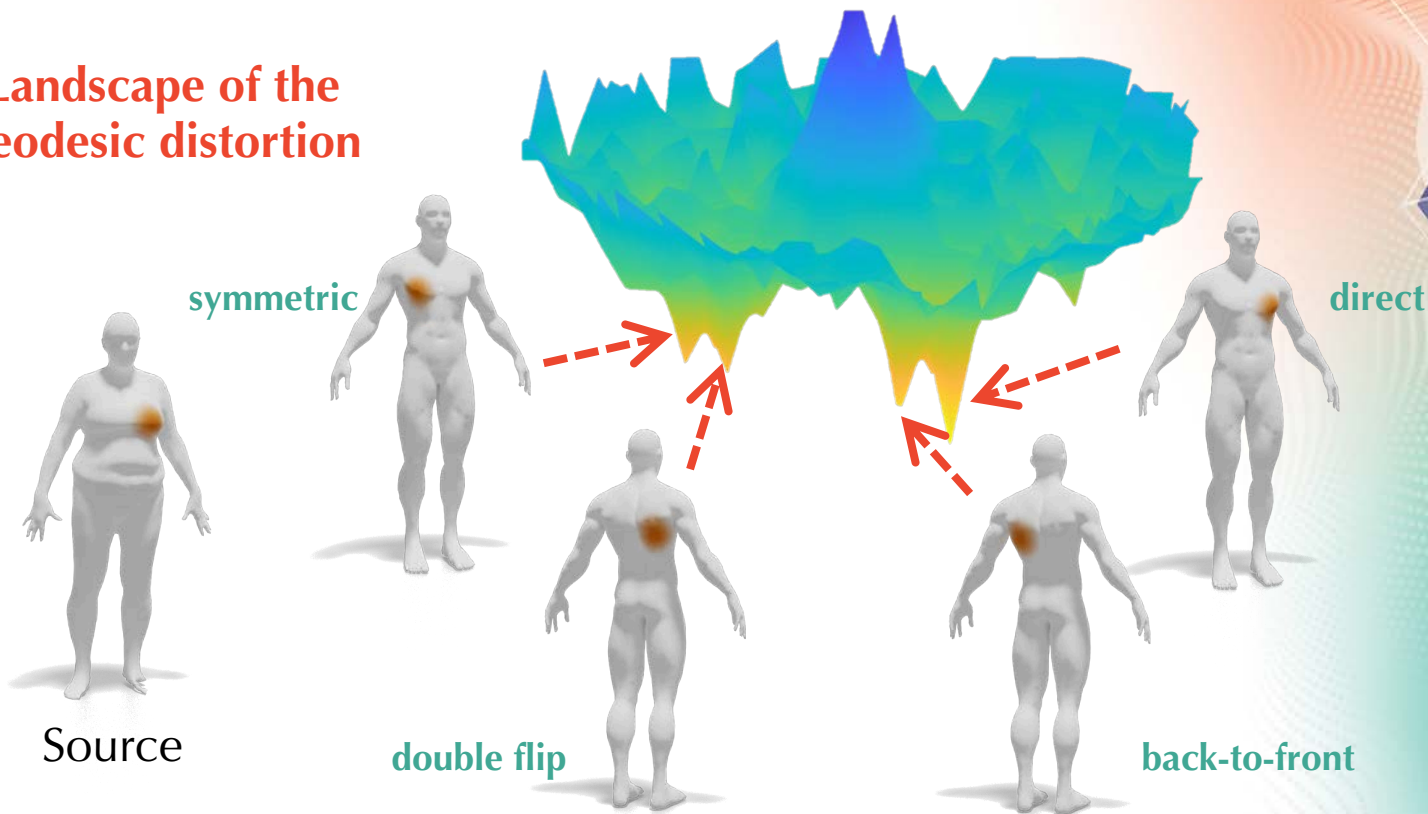


Source



Find all meaningful maps!

Landscape of the geodesic distortion



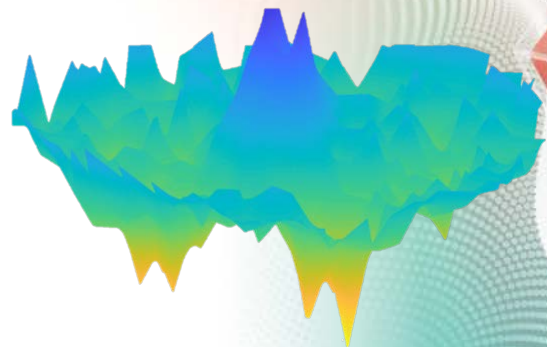
Find all meaningful maps!

Observations

- Intrinsic objectives: not discriminative w.r.t. self-symmetries
- Multiple ground-truth can exist!

Inspirations

- ~~Avoid symmetry flip during computation~~
- Find all meaningful maps and select later!
- Use **Functional Map** representation



Laplace-Beltrami Operator

Helmholtz Equation

$$\Delta_S f = \lambda f$$

Shape S



ϕ_1



$$0 = \lambda_1 \leq$$

ϕ_2



$$\lambda_2 \leq$$

ϕ_3



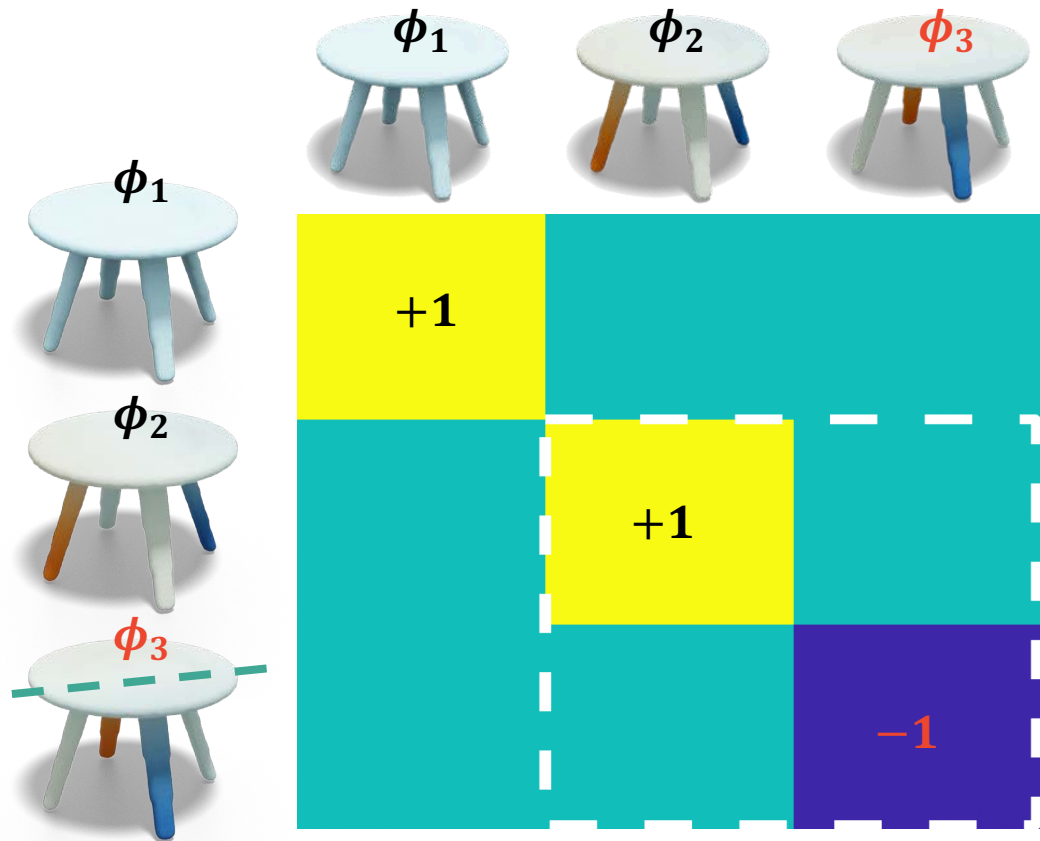
$$\lambda_3 \leq$$

ϕ_4



$$\lambda_4 \leq \dots$$

Functional map

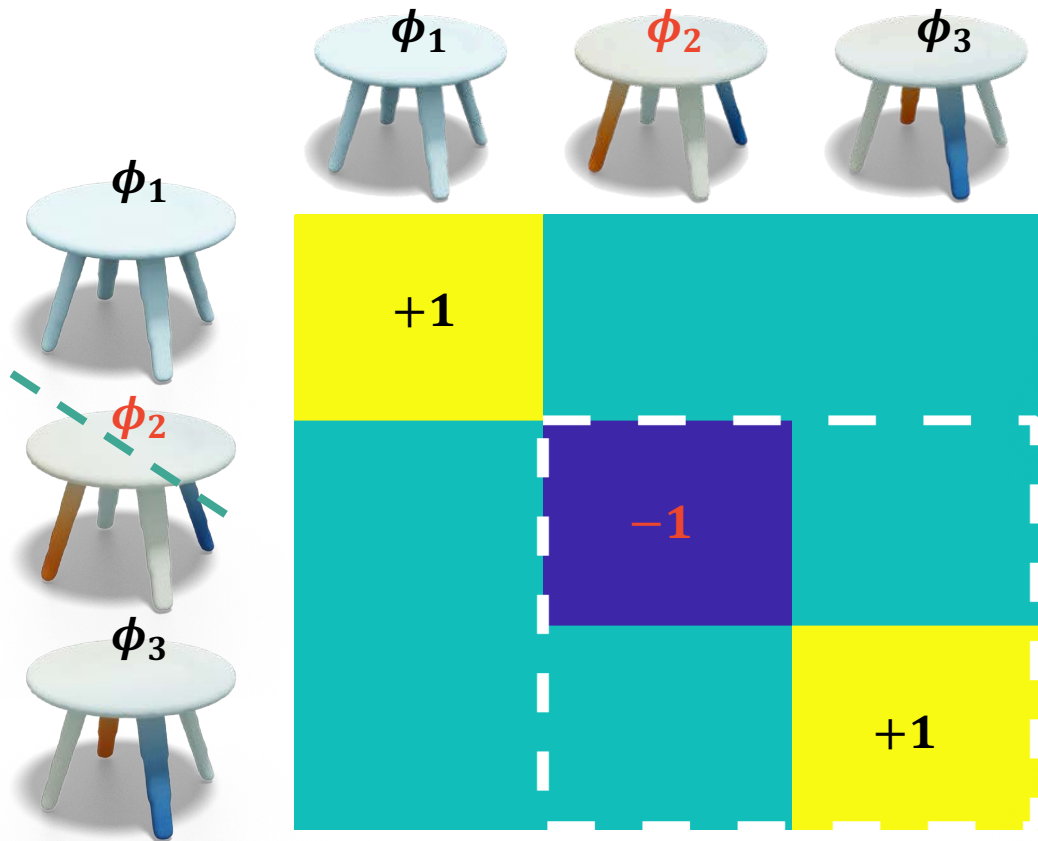


Source



Self-Map

Functional map

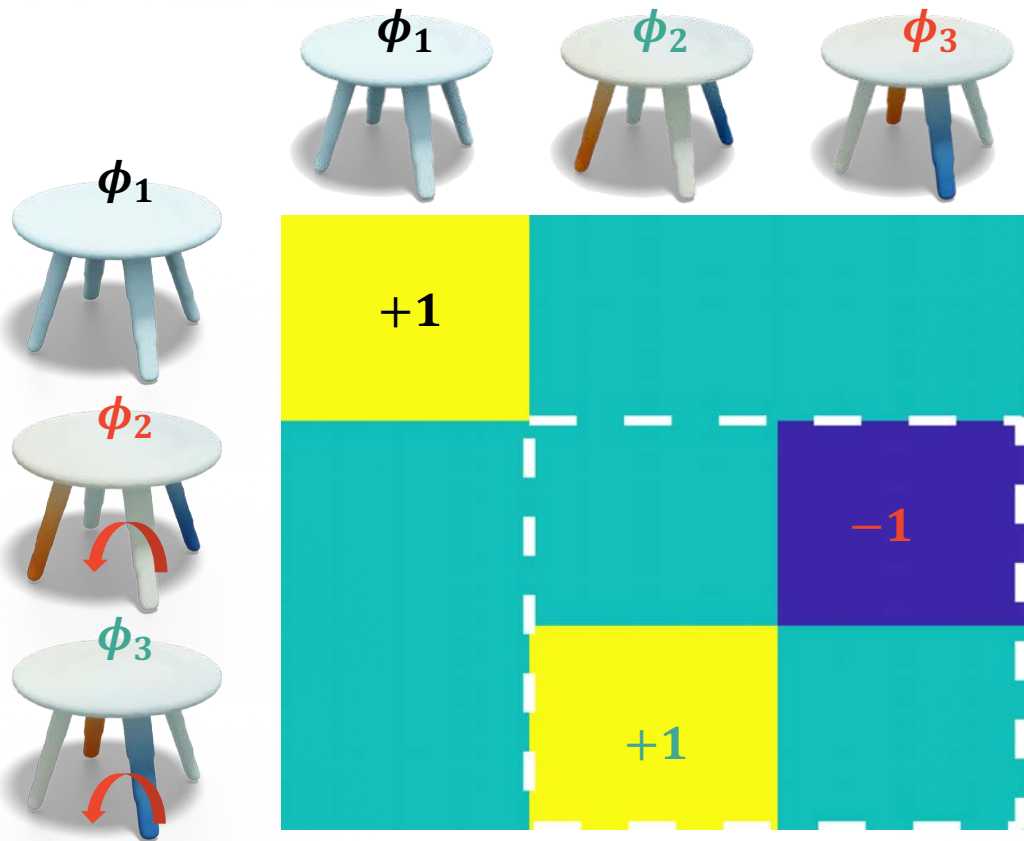


Source



Self-Map

Functional map



Source



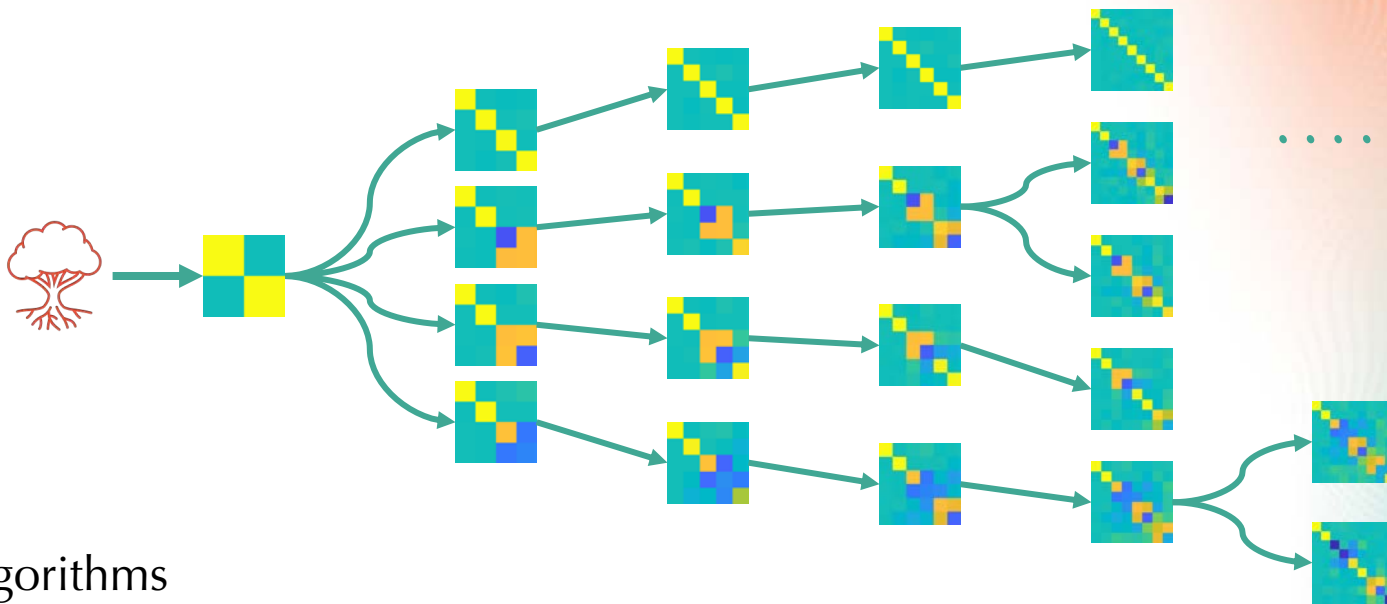
Self-Map

Functional map

Observations

- Intrinsic (global) symmetry information is encoded in the **spectral domain** of the shapes
- Pointwise maps can be organized in the spectral domain **along the frequencies**
- Use **F-norm between two functional maps** to approximate the distance between two pointwise maps

Map Tree Exploration



Algorithms

1. Functional map **expansion** rules
2. Map tree **exploration**

Map Tree Exploration

Contributions

- **Progressive exploration** of the map space
 - Start from the smallest functional map
 - Expand the functional map along the frequency domain
 - Only keep good & sufficiently different maps at each iteration

Contributions

- Output **multiple** maps, e.g.,
 - 4 maps for human v.s. human, human v.s. gorilla
 - 2 maps for animal shape pairs
 - ≥ 4 maps for man-made objects (table, knots, glasses, cup...)
- Previous work: a **single** map

Contributions

- **Bijjective ZoomOut** for map refinement
 - ZoomOut [Melzi et al 2019]: spectral refinement method that enforces the **orthogonality** of a functional map
 - We propose Bijjective ZoomOut that enforces both
 - **Orthogonality** of a single functional map
 - **Bijjectivity** of the functional maps from both sides
 - **Similar** computation complexity
 - **Better** accuracy

- Multi-solution shape matching
- Self-symmetry detection
- Non-rigid shape matching

Multi-solution shape matching

Source shape



Direct map



Symmetric



Back-to-front



Left-to-right &
Back-to-front



Multi-solution shape matching

Source shape



Left-to-right



Upside-down



Left-to-right &
Upside-down



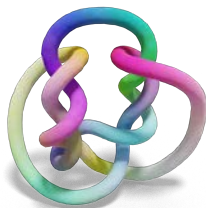
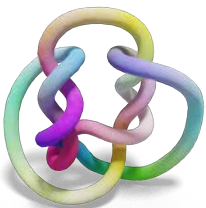
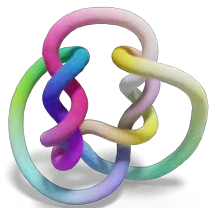
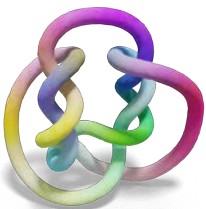
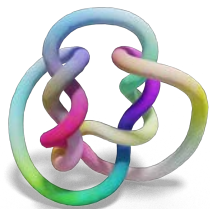
Multi-solution shape matching

Source

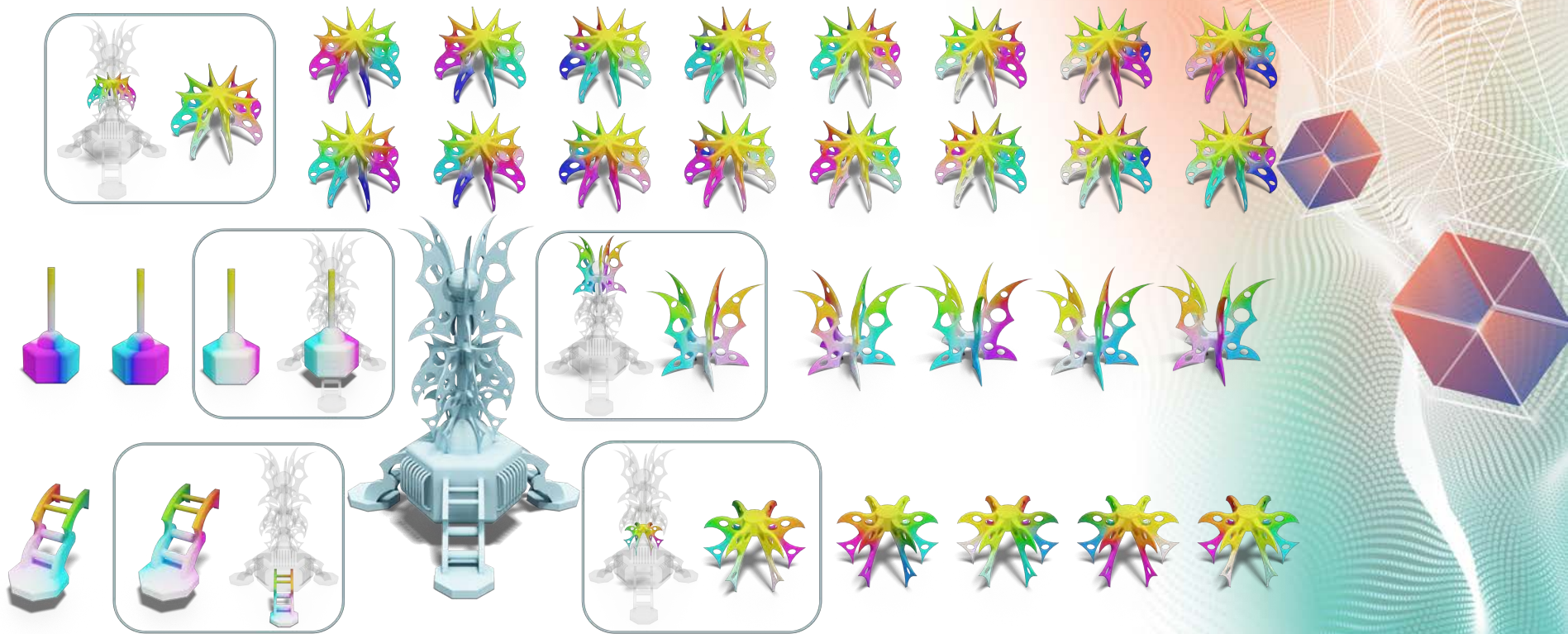
OrientRev

OrientRev +
ZoomOut

Ours



Multi-solution shape matching



Non-rigid shape matching

BIM

Orient (Ini)

ICP

PMF

BCICP

RHM

ZoomOut



Symmetry
ambiguity!



Source



Ours

Non-rigid shape matching

SHREC'19 challenge: state-of-the-art accuracy

Methods \ Measurement	Accuracy ($\times 10^{-3}$)	GeoDist ($\times 10^2$)	Dirichlet Energy	Conformal Distortion	Runtime (sec)
BIM	83.69	1.418	3.278	1.970	164
GroupRep	311.1	5.254	13.41	7.787	3.95
IntSymm	62.50	1.945	12.17	7.123	2.05
OrientRev (Ini)	137.2	4.682	22.07	12.69	0.52
Ini + ICP	108.9	3.604	10.49	6.235	8.33
Ini + PMF	119.4	2.444	15.98	9.605	425
Ini + RHM	118.7	4.166	7.352	4.369	28.7
Ini + BCICP	96.72	2.466	5.633	3.741	157
Ini + ZoomOut	80.30	2.858	6.601	3.838	6.58
MapTree - GT	39.62	1.512	3.763	0.949	65.2
MapTree - Auto	47.48	1.507	3.833	0.929	65.2

We also propose an **auto-selection** algorithm via **cycle-consistency!**

Self-symmetry detection

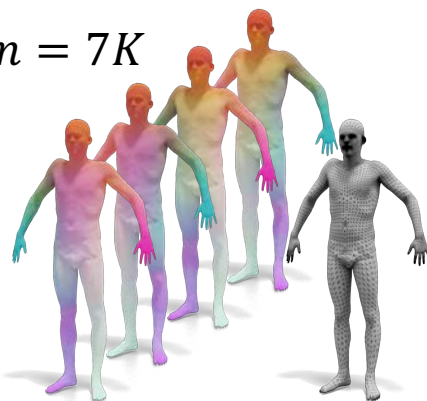


Robustness w.r.t. decimation

$n = 5K$



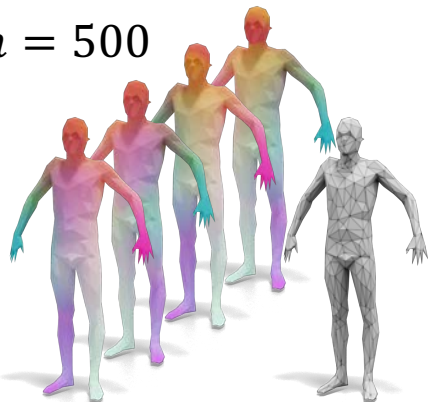
$n = 7K$



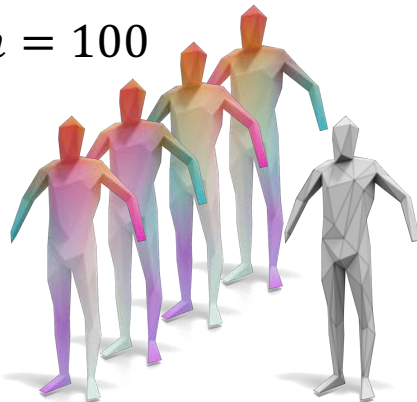
$n = 1K$



$n = 500$



$n = 100$



$n = 50$



Thank you for your attention 😊

MapTree:

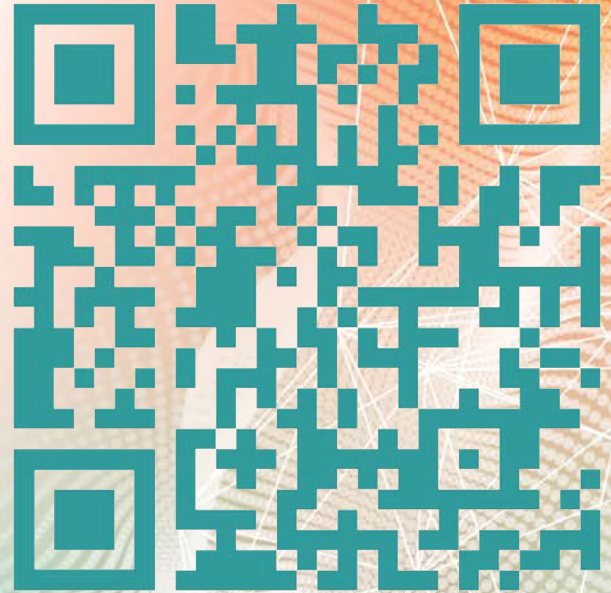
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Code:

https://github.com/llorz/SGA20_mapExplor